



Digital Development and Open Government

Evidence from Chilean Local Governments

Bastían González-Bustamante and **Diego Aguilar**

University of Oxford and Training Data Lab

✉ bastian.gonzalezbustamante@politics.ox.ac.uk

✉ diego.aguilar@training-datalab.com

Presentation at the Digital Democracy Workshop

Digital Society Initiative, University of Zurich

Zurich, October 27-28, 2022

Table of Contents

1. Introduction
2. Empirical Expectations
3. Empirical Strategy
4. Results
5. Discussion



Introduction

Our main question is: **What factors determine the development of open e-government in Chilean municipalities?**

We have constructed an index that combines elements of e-government with transparency indicators for all Chilean municipalities ($n = 345$) between 2019 and 2021. Then, we evaluated digital development at the local using **geospatial econometric models**.

Our empirical expectations focus on the first three dimensions related to: (i) **infrastructure**; (ii) characteristics of the population and, in particular, its **socioeconomic level**; and (iii) local **administrative capacity**.

Empirical Expectations

There is no consensus about the magnitude of the influence of infrastructure on digital development and transparency. [Sicáková-Beblavá et al. \(2016\)](#) found that it is not possible to confirm that greater Internet access leads to a higher level of government transparency. [Lowatcharin and Menifield \(2015\)](#) also reached a similar conclusion.

Even so, our empirical expectation is to find a positive relationship between infrastructure and the development of open e-government in Chilean municipalities ([González-Bustamante et al., 2020](#)).

- **Infrastructure Hypothesis.** The number of fixed Internet connections in a municipal district increases the municipality's development of open e-government.

In the case of the population's characteristics, the variables highlighted by the literature include age range and socioeconomic level. On the socioeconomic level, a study by [Dias and Costa \(2013\)](#), for example, concluded that the population's income level explains the demand for access to information.

We focus on the aggregate socioeconomic level measured inversely using monetary poverty indicators.

- **Socioeconomic Level Hypothesis.** A municipal district's level of monetary poverty reduces the municipality's development of open e-government.

The municipal government's own revenues should also be important (Dias and Costa, 2013; González-Bustamante et al., 2020).

We assert that the greater a municipal government's autonomous financing, the higher the level of open e-government will be. There will, therefore, be a difference with municipalities that do not have sufficient sources of revenues and, probably, have weaker organisational capacities.

- **Financial Resources Hypothesis.** A municipality's own revenues increase its development of open e-government.

We also consider the technical capabilities of municipal personnel as a determinant of the development of open e-government. [Chapman \(2017\)](#), for example, identifies a municipal government's professionalisation as explaining its level of adoption of e-government and innovation.

- **Administrative Capacity Hypothesis.** The level of professionalisation of municipal personnel increases the municipality's development of open e-government.

Empirical Strategy

We started by using the **e-services model** of [Esteves \(2005\)](#), which contains different items grouped into five dimensions, to evaluate government websites ([Fath-Allah et al., 2017](#)).

We applied the model to Chile's 345 municipalities. To this end, we carried out a binary measurement for each item and an aggregation, derived from the original formula of [Esteves \(2005\)](#), which takes into account the weights theoretically associated with the level of sophistication or digital maturity of each phase for the *i-th* municipalities:

$$Y_i = f_{1[i]} \times (0.25) + f_{2[i]} \times (0.50) + f_{3[i]} \times (0.75) + f_{4[i]} \times (1.00) + f_{5[i]} \times (1.25) \quad (1)$$

Municipal E-Services Model

| | Phase | Item |
|-------|-------------------|--|
| f_1 | Presence | Forms (downloadable documents) Municipal newsletter Online map |
| f_2 | Urban information | Urban map Transport information |
| f_3 | Interaction | Municipal social networks Municipal telephone Mobile |
| f_4 | Transaction | Online official procedures Follow-up Digital certificates Register of residents Online payment |
| f_5 | E-democracy | Citizen participation |

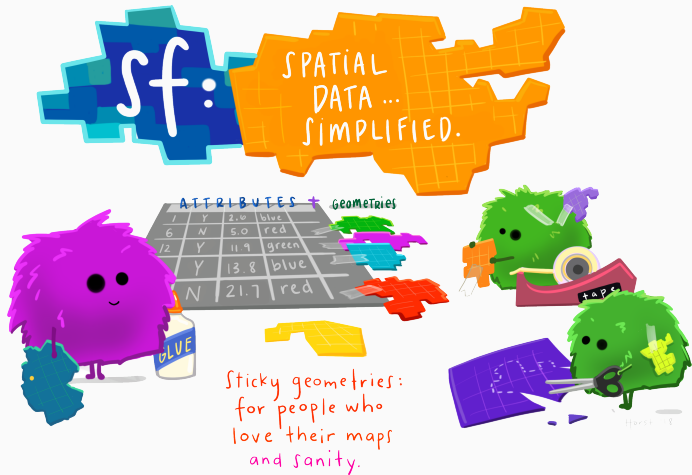
Some items in the original model, such as the presence of a **search engine** or the option of **personalising the interface** with the website, have been omitted so as to give priority to the incorporation of more important elements that are now of daily use, such as **digital social networks**.

We carried out measurements of municipal e-government, obtaining an estimate of the EGi. **We then expanded the EGi by incorporating transparency indicators.**

We used the requests for access to public information (**Freedom of Information Act, FOIA**) to construct an indicator of the average number of working days that each municipal government took to respond to requests.

Our aggregate OEGi has a Cronbach's α of 0.681 for all Chile's municipalities (95% CI [0.635, 0.719]). In the regions with over one million inhabitants, the indicator increases to 0.713 (95% CI [0.658, 0.758]). We carried out multiple imputations below 5% of cases.

Independent Variables



Independent Variables

We integrated the index with **geospatial** information and local government indicators compiled using the following open data and public information:

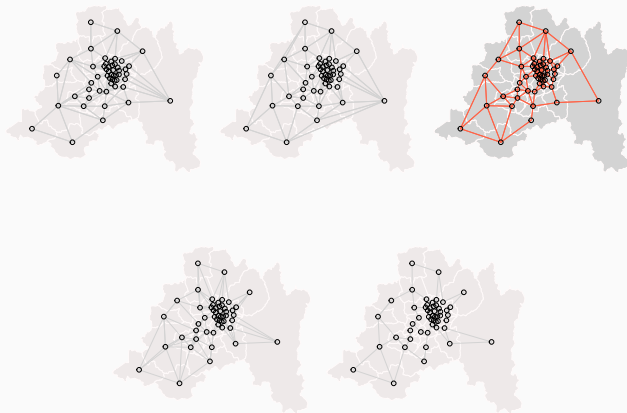
- Geospatial shapefiles ([IDE-Chile and SUBDERE, 2018, 2020](#))
- Number of fixed Internet connections ([SUBTEL, 2022](#))
- Rate of monetary poverty with a small area estimation (SAE) ([DOS-MDSF and ECLAC, 2021](#))
- Indicators of municipal budget with monetary correction and the professionalisation of municipal personnel based on the proportion holding a professional qualification ([SINIM, 2022](#))
- Estimated population of the municipal district ([INE, 2022](#))

We used **Moran's Index** (Moran's I), which is analogous to the Pearson coefficient for spatial units (Goodchild, 2008). We used one variant **under randomisation** and another with **Markov chain Monte Carlo** (MCMC) simulations for cross-validation.

Values close to zero indicate randomness in the spatial pattern, while values close to 1 and close to -1 indicate perfect correlation and perfect dispersion, respectively.

We worked with vector data from IDE-Chile and SUBDERE (2018, 2020) and spatial matrices based on the **Sphere of Influence (SOI) model**, derived from the Delaunay triangulation (Bivand et al., 2013).

Types of Contiguity Matrices



(i) Queen-style contiguity matrix; (ii) matrix with Delaunay triangulation; (iii) matrix with SOI model, highlighted in colour; (iv) matrix based on neighbours by distance with $k = 4$; and (v) matrix based on neighbours by distance with $k = 2$.

We then used **OLS models**. Our baseline model considers *i-th* municipalities ($n = 345$) and regresses our index $Y_i = \text{EOGi}_{[i]}$ on the number of **fixed Internet connections** $X_{1[i]}$. As FE, we then incorporated the municipal district's **estimated population** and **population density**, before also incorporating the municipal district's **rate of monetary poverty** $X_{2[i]}$ and, finally, each municipal government's **own permanent revenues** $X_{3[i]}$ and the **rate of professionalisation of personnel** $X_{4[i]}$. We used some logarithmic transformations.

$$Y_i = \alpha + \beta_1 \log(X_{1[i]}) + \beta_2 X_{2[i]} + \beta_3 \log(X_{3[i]}) + \beta_4 X_{4[i]} + \gamma_1 \log(\text{pop}_i) + \gamma_2 \log(\text{density}_i) + \varepsilon_i \quad (2)$$

With the model's residuals, spatial autocorrelation can be evaluated statistically with Moran's I. If the test is statistically significant at 95% confidence, the residuals are spatially grouped, and it is appropriate to fit spatial econometric models.

The options include **spatial autoregressive (SAR) models** in which the ρW parameter measures the spatial autocorrelation of the dependent variable. Considering our vector of the j -th main independent variables ($j = 4$), the equation is as follows, applying the transformation of the vector's variables when necessary:

$$Y_i = \rho W Y_i + \alpha + \sum_{j=1}^4 \beta_j X_{j[i]} + \gamma_1 \log(pop_i) + \gamma_2 \log(density_i) + \varepsilon_i \quad (3)$$

Another option corresponds to **spatial error models (SEM)**, where the parameter λWu measures the spatial dependence of the errors for a latent continuity variable u .

$$Y_i = \alpha + \sum_{j=1}^4 \beta_j X_{j[i]} + \gamma_1 \log(pop_i) + \gamma_2 \log(density_i) + \lambda Wu_i + \varepsilon_i \quad (4)$$

Results

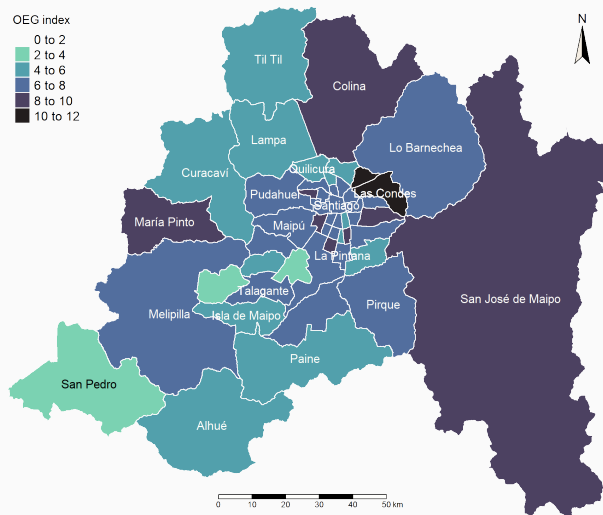
Georeferencing and Spatial Autocorrelation

We detected patterns of **weak spatial dependence at the national level**, with a 95% statistical significance and a Moran's I of 0.183 under randomisation ($p \leq 0.001$) and for the MCMC simulations ($p \leq 0.001$). Only three regions show significant **patterns of autocorrelation** (Tarapacá, Araucanía and Los Lagos Regions).

The Metropolitan Region, the most populous of all Chile's regions, has the country's highest average index: 6.722 out of a maximum possible of 12.5 points. However, the OEGi's spatial distribution does not show significant patterns at the regional level (Moran's I $p = 0.105$, MCMC $p = 0.110$).

Another region that stands out for its high average is the Valparaíso Region, the third most populous, while the areas in the extreme north and south of the country have comparatively lower indices.

Open E-Government Index in the Metropolitan Region



Determinants of Open E-Government at the Local Level

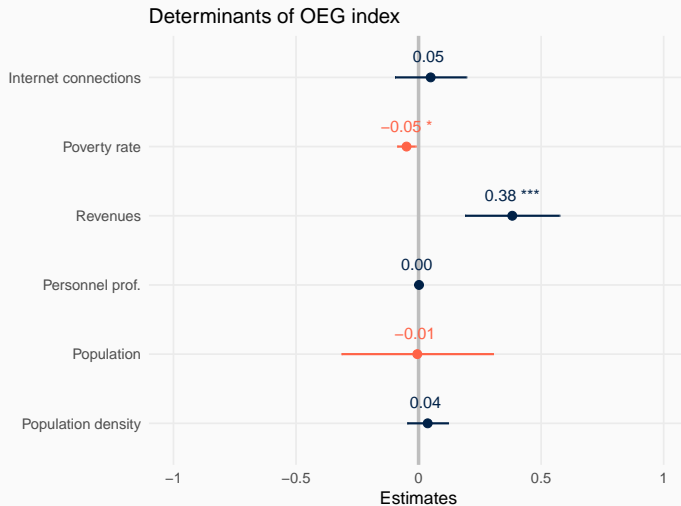
When we incorporated, in model III, the municipal district's monetary poverty rate, the number of **Internet connections ceases to be statistically significant** ($p = 0.307$).

We fitted SAR and SEM models to adequately measure connections' effect on the OEGi.

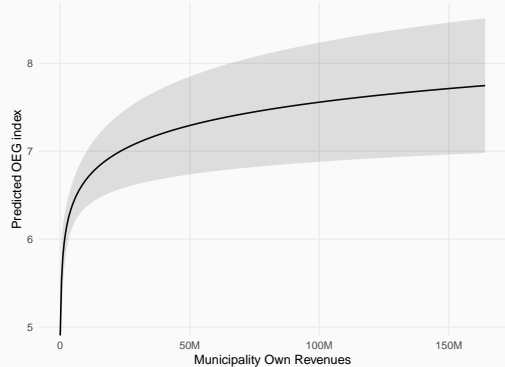
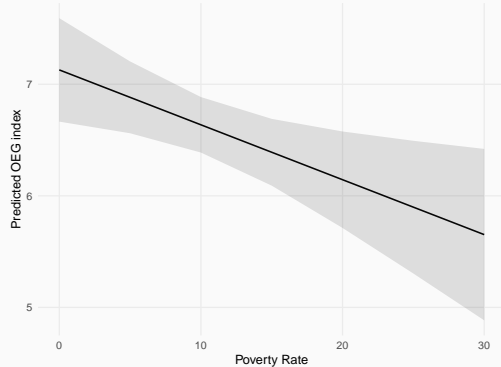
| | Model I | Model II | Model III | Model IV | Model V |
|-------------------------------------|---------------------|---------------------|----------------------|----------------------|---------------------|
| Internet connections (<i>log</i>) | 0.285*** (0.027) | 0.193*** (0.073) | 0.077 (0.075) | 0.048 (0.074) | 0.049 (0.074) |
| Poverty rate | | | -0.083*** (0.017) | -0.049*** (0.019) | -0.049** (0.019) |
| Revenues (<i>log</i>) | | | | 0.382*** (0.098) | 0.382*** (0.098) |
| Personnel professionalisation rate | | | | | 0.002 (0.006) |
| Constant | 3.773*** (0.206) | 2.827*** (0.956) | 3.511*** (0.937) | 0.528 (1.193) | 0.425 (1.241) |
| Population (<i>log</i>) | No | Yes | Yes | Yes | Yes |
| Population density (<i>log</i>) | No | Yes | Yes | Yes | Yes |
| Estimation method | OLS | OLS | OLS | OLS | OLS |
| VIF | 1.324 | 1.331 | 1.421 | 1.486 | 1.486 |
| B-P/Cook-Weisberg | 0.031 | 0.001 | 1.364 | 1.443 | 1.437 |
| Shapiro-Wilk | 0.991** | 0.991** | 0.995 | 0.994 | 0.994 |
| Moran's I | 0.081** | 0.082** | 0.033 | 0.016 | 0.018 |
| <i>N</i> | 345 | 345 | 345 | 345 | 345 |
| <i>R</i> ² | 0.245 | 0.249 | 0.296 | 0.327 | 0.327 |
| Adj. <i>R</i> ² | 0.242 | 0.242 | 0.288 | 0.317 | 0.315 |

* $p \leq 0.1$; ** $p \leq 0.05$; *** $p \leq 0.01$

Determinants of Open E-Government at the Local Level



Marginal Effects on Open E-Government at the Local Level in Chile



We incorporated a binary variable that measures the existence of **municipal protocols on citizen participation**. None of our tests showed it to be significant.

We also alternated the municipal government's revenues variable with measurements that reflect the level of **local tax collection** and the **municipal government's financial dependence**. Only this last variable was significant, with a negative coefficient at 90% confidence ($p = 0.055$), which is consistent with financial dependence as an aspect of weakness in organisational capacity.

Discussion

We present a **novel open e-government index** based on the **e-services model** of Esteves (2005) and **transparency indicators**. At the national level, a slight spatial autocorrelation is detected but, at the subnational level, only the Tarapacá, Araucanía and Los Lagos Regions show clustering.

Our evidence allows us to **accept** the **Socioeconomic Level Hypothesis** and the **Financial Resources Hypothesis** since a municipal district's monetary poverty rate causes decreases in our index while the level of the municipal government's own resources increases it. This is consistent with the literature on open e-government and digital development.

On the other hand, we **reject** the **Infrastructure Hypothesis** and the **Administrative Capacity Hypothesis**. The number of fixed Internet connections in a municipal district is significant in the absence of other factors, and the econometric modelling reveals patterns of spatial autocorrelation.

The index we have presented constitutes a **theoretical contribution** that implies dialogue between e-government and open government concepts. It is, moreover, a **methodological and empirical** contribution that can serve as a starting point for future research focusing on problems of non-random selection with observational data at subnational levels.

References

References I

- Bivand, R. S., Pebesma, E., and Gómez-Rubio, V. (2013). *Applied Spatial Data Analysis with R*. Springer, New York.
- Chapman, S. J. (2017). What's in a website? E-Government Scores and Municipal Characteristics. *Public Administration Quarterly*, 41(2):360–385.
- Dias, G. P. and Costa, M. (2013). Significant socio-economic factors for local e-government development in Portugal. *Electronic Government, An International Journal*, 10(3/4):284–309.
- DOS-MDSF and ECLAC (2021). Estimaciones Comunes de Pobreza por ingresos en Chile Mediante Métodos de Estimación en Áreas Pequeñas. Working Paper, División Observatorio Social del Ministerio de Desarrollo Social y Familia (DOS-MDSF, Social Observatory Division of the Ministry of Social Development and the Family), Gobierno de Chile and UN Economic Commission for Latin America and the Caribbean (ECLAC).
- Esteves, J. (2005). Análisis del desarrollo del gobierno electrónico municipal en España. Working Paper WPE05-32, IE - Reinventing Higher Education. IE Working Paper.

References II

- Fath-Allah, A., Cheikhi, L., Idri, A., and Al-Qutaish, R. (2017). A Best Practice Based E-Government Portals' Maturity Model - A Case Study. In *IEEE International Conference on Computer and Information Technology*, Helsinki.
- González-Bustamante, B., Carvajal, A., and González, A. (2020). Determinantes del gobierno electrónico en las municipalidades: Evidencia del caso chileno. *Gestión y Política Pública*, XXIX(1):97–129.
- Goodchild, M. (2008). Spatial Autocorrelation. In Kemp, K., editor, *Encyclopedia of Geographic Information Science*. SAGE Publications, Thousand Oaks.
- IDE-Chile and SUBDERE (2018). Planilla Códigos Únicos Territoriales (CUT). Dataset, Infraestructura de Datos Geoespaciales (IDE-Chile, Infrastructure of Geospatial Data of Chile), Chile and Subsecretaría de Desarrollo Regional y Administrativo (SUBDERE, Undersecretariat for Regional Development), Gobierno de Chile. Available at <https://www.ide.cl>.
- IDE-Chile and SUBDERE (2020). División Política Administrativa 2020. Dataset, Infraestructura de Datos Geoespaciales (IDE-Chile, Infrastructure of Geospatial Data of Chile), Chile and Subsecretaría de Desarrollo Regional y Administrativo (SUBDERE, Undersecretariat for Regional Development), Gobierno de Chile. Available at <https://www.ide.cl>.

References III

- INE (2022). Proyecciones de Población 2002-2020. Dataset, Instituto Nacional de Estadísticas (INE, National Institute of Statistics), Chile.
- Lowatcharin, G. and Menifield, C. E. (2015). Determinants of Internet-enabled Transparency at the Local Level. *State and Local Government Review*, 47(2):102–115.
- Sicáková-Beblavá, E., Kollárik, M., and Sloboda, M. (2016). Exploring the determinants of transparency of Slovak municipalities. *Network of Institutes and Schools of Public Administration in Central and Eastern Europe. The NISPAcee Journal of Public Administration and Policy*, 9(2):121–145.
- SINIM (2022). Sistema Nacional de Información Municipal (SINIM). Dataset, Subsecretaría de Desarrollo Regional y Administrativo (SUBDERE, Undersecretariat for Regional Development), Gobierno de Chile. Available at <http://datos.sinim.gov.cl>.
- SUBTEL (2022). Estadísticas servicio de acceso a Internet fija: Conexiones fijas por región y comuna. Dataset, Subsecretaría de Telecomunicaciones (SUBTEL, Undersecretariat for Telecommunications), Gobierno de Chile.

Thank you very much!



Presentation compiled with \LaTeX and some 

 Download the latest version from [GitHub](#)